



VAIDS-DSML exercise block 9

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simple linear regression

- problem

- 1599 samples of wines
- 11 features
- 1 target: quality
- 1 feature as linear regressor: alcohol

- solution

- use `sklearn.linear_model`
- construct model manually
- compare ols parameters (slope and intercept)

simple linear regression

function for linear regression

$$y_i = \beta_0 + \beta_i x_i + \varepsilon_1$$

as matrices

$$\mathbf{y} = \mathbf{x}\boldsymbol{\beta} + \boldsymbol{\varepsilon}$$

dimensions of those matrices in simple linear regression

$$(n \times 1) = (n \times 2)(2 \times 1) + (n \times 1)$$

coefficients

$$\boldsymbol{\beta} = \begin{bmatrix} \beta_0 \\ \beta_1 \end{bmatrix}$$

estimate for $\boldsymbol{\beta}$

$$\mathbf{b} = \begin{bmatrix} b_0 \\ b_1 \end{bmatrix}$$

prediction \hat{y}

$$\hat{\mathbf{y}} = \mathbf{x}\mathbf{b}$$

simple linear regression

error e

$$e = y - \hat{y}$$

or

$$e = y - xb$$

note dimensions

$$(n \times 1) = (n \times 1) - (n \times 1)$$

remember the goal is to minimize the error:

$$\min \sum_i^n e_i^2$$

or

$$\min e^T e$$

or

$$\min (y - xb)^T (y - xb)$$

note dimensions

$$(1 \times n)(n \times 1) = (1 \times 1)$$

simple linear regression

we want to find the b which minimizes the error, so we take the partial derivative with regard to b and set it to 0

$$\frac{\partial}{\partial b}(\mathbf{y} - \mathbf{x}b)^T (\mathbf{y} - \mathbf{x}b) = 0$$

draw the transposition into the parentheses and remember to reverse the terms if you transpose a product

$$\frac{\partial}{\partial b}(\mathbf{y}^T - \mathbf{b}^T \mathbf{x}^T)(\mathbf{y} - \mathbf{x}b) = 0$$

aus-x-en

$$\frac{\partial}{\partial b} \mathbf{y}^T \mathbf{y} - \mathbf{y}^T \mathbf{x}b - \mathbf{b}^T \mathbf{x}^T \mathbf{y} + \mathbf{b}^T \mathbf{x}^T \mathbf{x}b = 0$$

simple linear regression manually

$$\frac{\partial}{\partial \mathbf{b}} \mathbf{y}^T \mathbf{y} - \mathbf{y}^T \mathbf{x} \mathbf{b} - \mathbf{b}^T \mathbf{x}^T \mathbf{y} + \mathbf{b}^T \mathbf{x}^T \mathbf{x} \mathbf{b} = 0$$

note the dimensions of only this part of the derivative

$$-\mathbf{y}^T \mathbf{x} \mathbf{b}$$

$$(1 \times n)(n \times 2)(2 \times 1)$$

$$(1 \times n)(n \times 1)$$

$$(1 \times 1)$$

so it is a scalar

and we have that two times in our term

$$-\mathbf{y}^T \mathbf{x} \mathbf{b} - \mathbf{b}^T \mathbf{x}^T \mathbf{y}$$

$$-2\mathbf{b}^T \mathbf{x}^T \mathbf{y}$$

so the whole term looks like this now

$$\frac{\partial}{\partial \mathbf{b}} \mathbf{y}^T \mathbf{y} - 2\mathbf{b}^T \mathbf{x}^T \mathbf{y} + \mathbf{b}^T \mathbf{x}^T \mathbf{x} \mathbf{b} = \mathbf{0}$$

simple linear regression

now derive: terms with no b disappear, terms with a b loose the b , terms with b^2 become $2b$

$$-2\mathbf{x}^T \mathbf{y} + 2(\mathbf{x}^T \mathbf{x})\mathbf{b} = \mathbf{0}$$

solve for b (the osl parameter estimates)

$$2(\mathbf{x}^T \mathbf{x})\mathbf{b} = 2\mathbf{x}^T \mathbf{y}$$

$$(\mathbf{x}^T \mathbf{x})\mathbf{b} = \mathbf{x}^T \mathbf{y}$$

$$(\mathbf{x}^T \mathbf{x})^{-1}(\mathbf{x}^T \mathbf{x})\mathbf{b} = (\mathbf{x}^T \mathbf{x})^{-1}(\mathbf{x}^T \mathbf{y})$$

$$(\mathbf{x}^T \mathbf{x})^{-1}(\mathbf{x}^T \mathbf{x}) = 1$$

$$\mathbf{b} = (\mathbf{x}^T \mathbf{x})^{-1}(\mathbf{x}^T \mathbf{y})$$

check the dimensions one last time

$$(2 \times 1) = (2 \times n)(n \times 2)(2 \times n)(n \times 1)$$

$$(2 \times 1) = (2 \times 2)(2 \times 1)$$

same as

$$\hat{\mathbf{y}} = \mathbf{x}\mathbf{b}$$

yay, checks out :)